

1. An Example of Bayesian Parameter Estimation

One of the easiest applications of Bayesian analysis is the estimation of the parameters of a statistical distribution. To see how it is done let us deal with a simple example: the exponential distribution

- a) What is the exponential distribution of parameter λ ?

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \geq 0 \\ 0, & x < 0 \end{cases}$$

- b) Write the general expression for likelihood of the sample vector $X=(x_1, \dots, x_N)$

$$L(D) = \prod_{i=1}^n \lambda e^{-\lambda x_i} = \lambda^n e^{-\lambda \sum_{i=1}^n x_i}$$

- c) It is customary to assign to scale parameters such as λ a prior distribution $P(\lambda) = \frac{1}{\lambda}$. Write the posterior of λ

$$P(\lambda | x) = \frac{P(x | \lambda)P(\lambda)}{P(x)}$$

- d) Find the general expression of the maximum of the posterior

$$\begin{aligned} P(\lambda | x) &= \lambda^{n-1} e^{-\lambda \sum_{i=1}^{n-1} x_i} \\ \ln P(\lambda | x) &= (n-1) \ln \lambda - \lambda \sum_{i=1}^{n-1} x_i \\ \frac{d}{d\lambda} \ln P(\lambda | x) &= (n-1) \frac{1}{\lambda} - \sum_{i=1}^{n-1} x_i, \\ \text{at which its maximum occurs at } \frac{d}{d\lambda} \ln P(\lambda | x) &= 0 \end{aligned}$$

$$\frac{1}{\lambda} = \frac{1}{(n-1)} \sum_{i=1}^{n-1} x_i$$

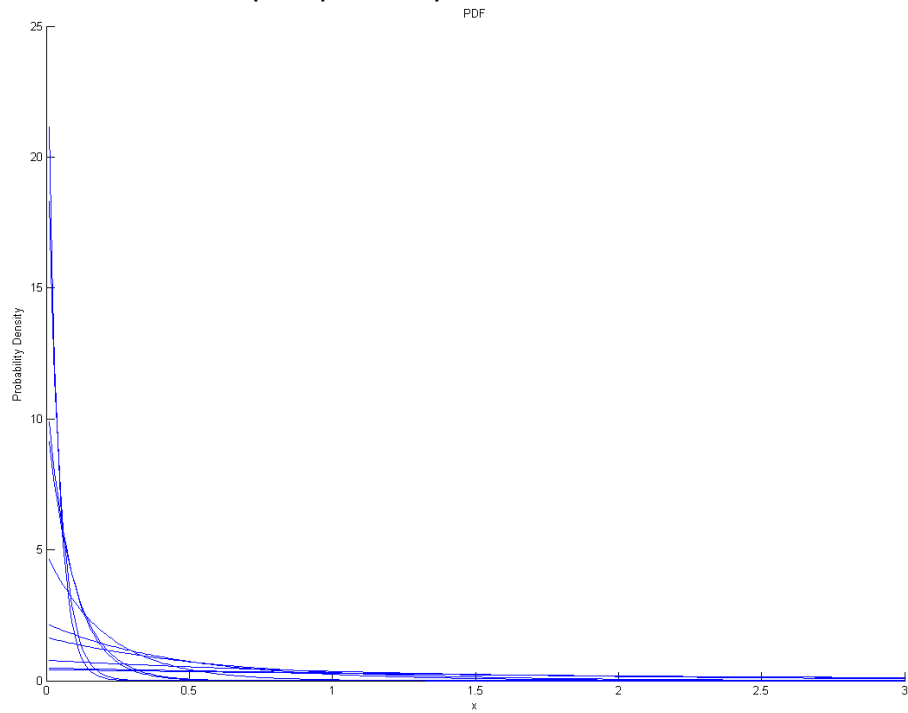
- e) **Bonus Question:** $P(\lambda)=1/\lambda$ is not normalised... What can be done about it?

$$\text{To normalise, } P(\lambda) = \frac{1/\lambda}{\int_0^\infty \lambda d\lambda}$$

- f) Draw 10 samples x_1, \dots, x_{10} from the exponential distribution $\lambda = 1$

0.2049	0.0989	2.0637	0.0906	0.4583
2.3275	1.2783	0.6035	0.0434	0.0357

g) For these 10 samples plot the posterior of λ



h) Calculate its maximum, its expected value and its standard deviation